

Traffic Camera Dangerous Driver Detection TCD3™

Contextually Aware Heuristic Feature & OFA Density-Based Computer Vision with
Movement Machine Learning Analysis of Live Streaming Traffic Camera Footage to
Identify Anomalous & Dangerous Driving

Outline

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- >Computer Vision
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Motivation

- >Summer Internship Project in Computer Science
- >Sensor Data Processing
- >Machine Learning

>Drunk/Distracted Driving is a serious problem

- >Personal Connection
- >Distracted Driving
 - >660,000 distracted drivers
 - >More than \$50,000,000 per year
- >Drunk Driving
 - >10.3 Million Drunk Drivers
 - >Requires human-in-the-loop
 - >More than \$40,000,000 per year



Technical Goal

- > Input
 - > Receive real time traffic camera data from surveillance, red light and traffic cameras
- > Process data in real time
- > Apply Computer Vision techniques to detect cars
- > Create Intelligent System that gets smarter to build an accurate model of normal driving for anomaly detection
- > Test for low quality footage
- > Optimize for high quality footage
- > Output
 - > 10 second video clips
 - > Sent to Police Car's Laptops
 - > Showing video of suspected distracted drivers



Language & Environment

- > C++
 - > Low-level access
 - > Matrix manipulation
- > Prototyped with Octave
- > 2015 Project
- >Redesigned with OpenCV
 - >Developed by Intel
 - >Large open-source community
 - >Low level access compared to other libraries

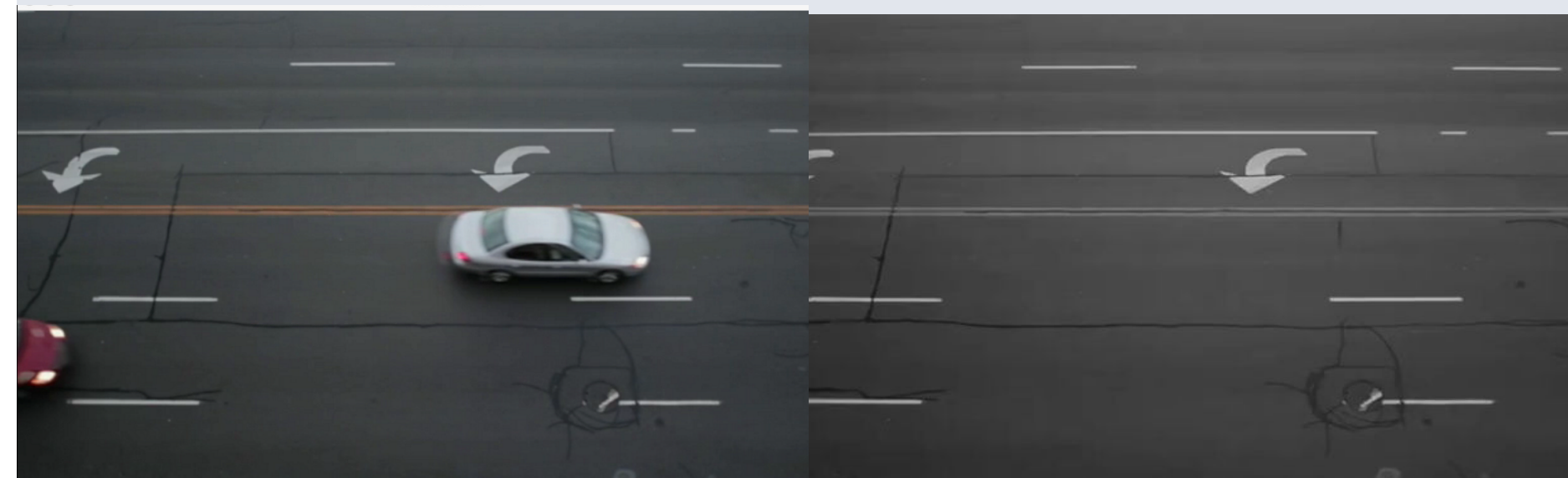
Computer Vision

- > Heuristic
 - > Series of practical steps to effectively analyze images
- > Background Subtraction
- > Background Generation
 - > Statistical Techniques
- > Optical Flow Techniques
- > Sliding Window Neighbor Detector
 - > Refining Data
- > Voting system to get clean detects
- > Return center point guess for cars for all methods
- > ~.59 Seconds per Frame
- > Accuracy of 92%

Vidur Prasad | Dayton Regional STEM School | Dayton, OH

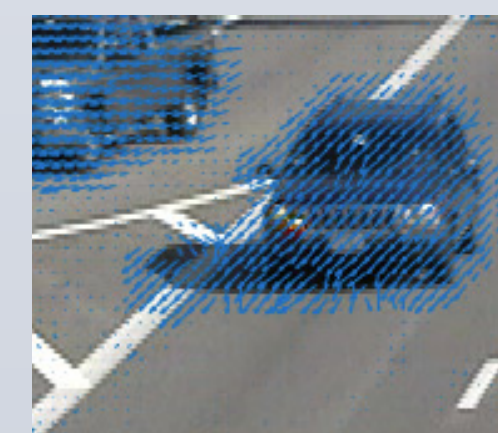
Background Model

- > Background Subtraction
 - > Find difference in frames
- > Update models to deal with moving sunlight
- > Background Median Model
 - > Median of Pixel Values
 - > Quick approach
 - > ~.14 seconds per frame
 - > Inaccurate
 - > 57% accuracy
- > Gaussian Mixture Model
- > MOG Gaussian Approach
 - > Lighting Adjustments
 - > Shadow Removal
 - > Regular Movement
- > ViBe
 - > Implements Probability Density Function
 - > Initial Training Period of 500 Frames
 - > Actively Updating
 - > Highest accuracy of methods



Optical Flow Analysis

- > Quantify motion in video
- > Determine the delta of pixels frame over frame
- > Farneback Model
- > Vectors returned quantifying movement
- > Irregular Object Detection
- > Future use
- > Supplement Background
 - > Diverse car colors have no effect
 - > Even small cars are picked up by OFA
 - > OFA can detect even 1 pixel movement



Raw Frame

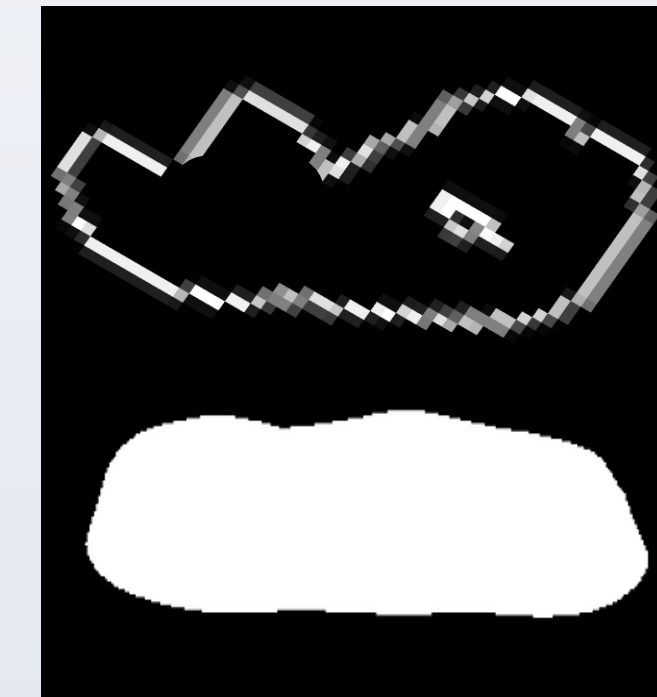


Background Image



Sliding Window Neighbor Detector

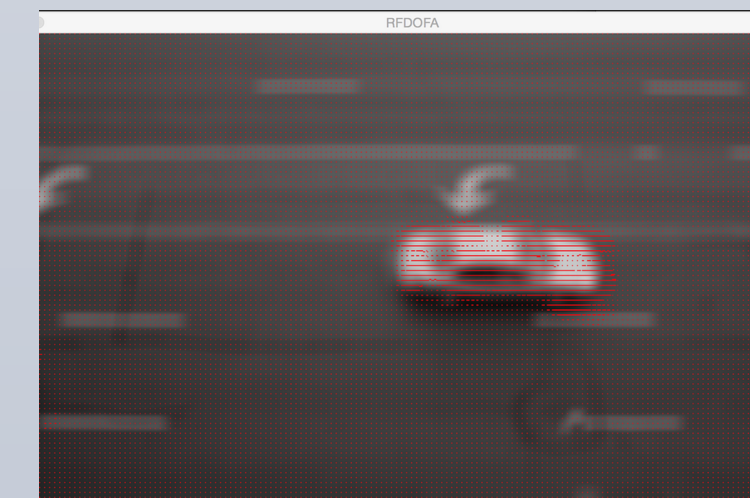
- > Morphology is Industry Standard
 - > Used to Refine Binary Image
- > Address Noise
- > Fix Errors in Binary Image
- > Recursive Definition
- > Density analysis
- > Process
 - > Create Window on Sensor Output
 - > Slide Window across Binary Image
 - > If Window is filled to threshold, then completely fill Window
 - > If Window is not filled to threshold, completely empty Window
 - > Increase size of window for multiple passes
- > Increases speed over industry by 62%
- > Reduces inaccurate detects by 47%



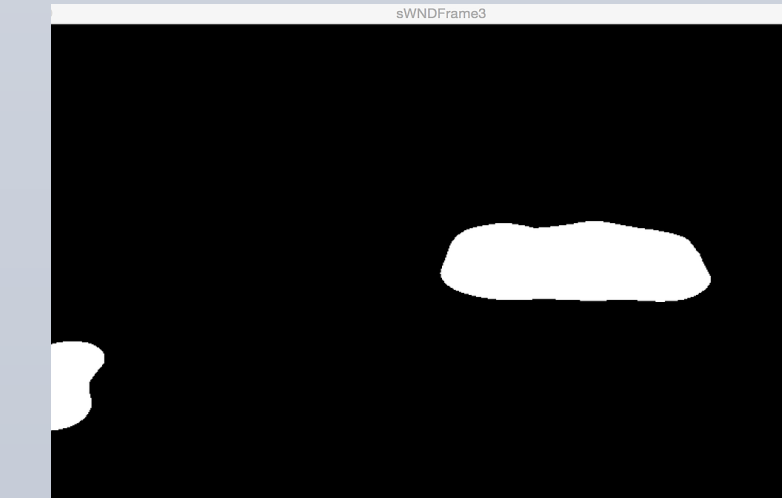
Machine Learning

- > Historical Analysis
- > Active Analysis
 - > "Duck on the road"
- > Matching Algorithm
 - > K Means
- > Learning Area Sector Model
- > Car Movement Properties
- > System Paradigm
 - > Supervised Algorithms
 - > Unsupervised Nature
- > Require violations in more than one properties
- > ~.43 Seconds per Frame
- > Accuracy of 83%

Optical Flow Analysis



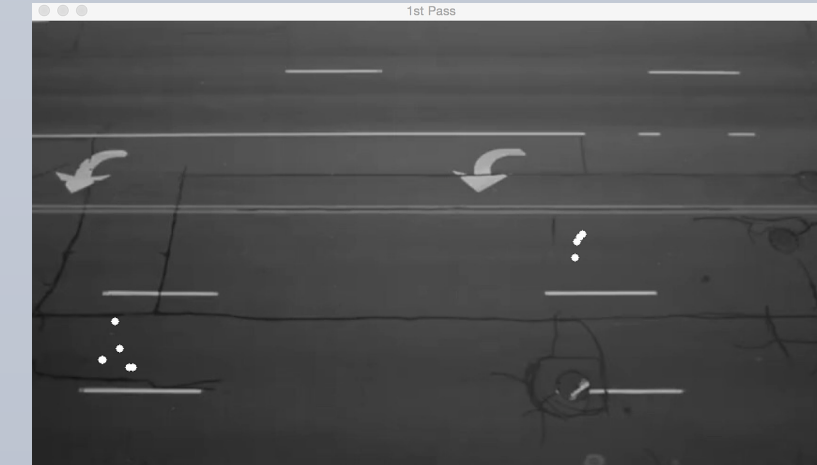
Sliding Window Neighbor Detector



Final Detection



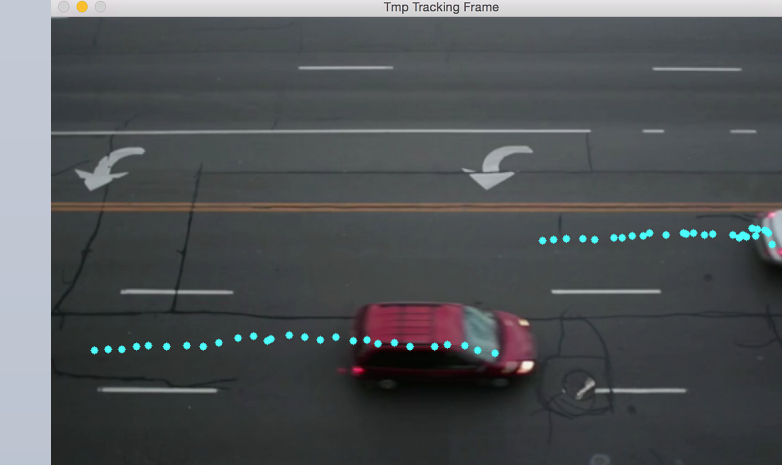
Car Detects



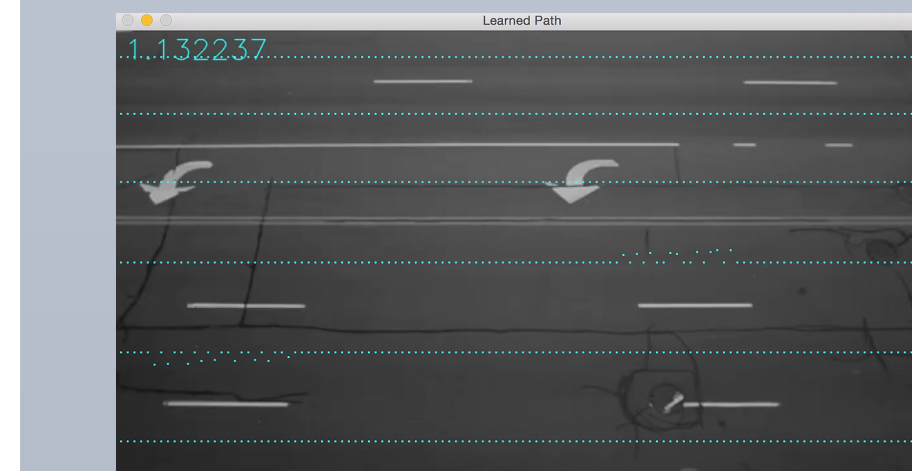
Refined Detects



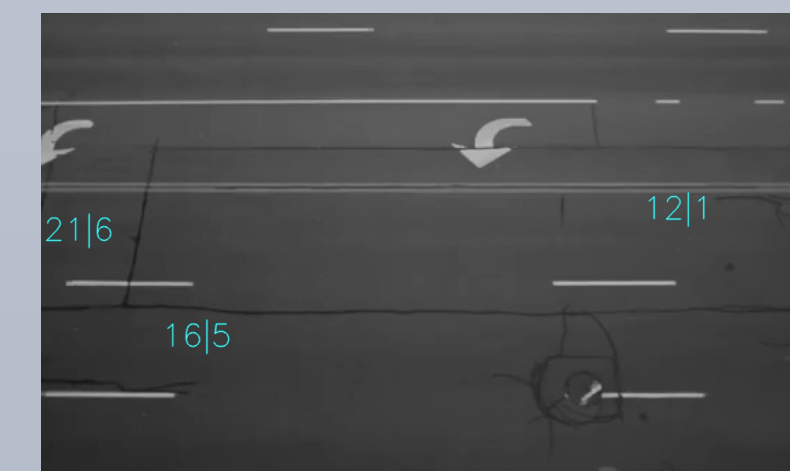
Track Detects



Learning Area Sector Model



Property Analysis



Anomaly Detection



Learning Area Sector Model

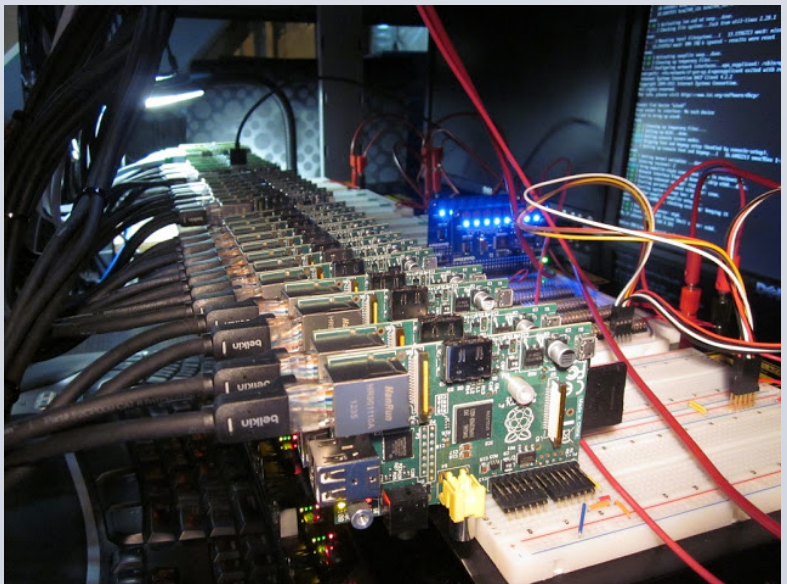
- > Learn area properties
 - > Complicated lanes
- > Develop sectors for discrete paths
- > Measure deviance from lane
 - > Learn average displacement
- > Measure average speed at points
- > Measure average acceleration at points
- > Measure angle at points
- > Utilize safety factor to determine acceptable range
- > Open Data Structure
 - > API
 - > Adapts to properties
- > Track lane entry departure
 - > Swerving

Property Learning

- > Learn individual car properties for LASM points
- > Hard Threshold for Speed Limits
- > Learned Threshold
- > Flexibility
 - > Alter allowed error with safety
- > Tag with weather and time
 - > Adjust models for accuracy
- > Sleep when speed drops below threshold
- > Learned Properties
 - > Speed
 - > Acceleration
 - > Direction
 - > Angle
- > 2 week training period

Distributed Computing

- > 1.02 seconds per frame
- > Multi-threading critical for performance
 - > Computer Vision Processes
 - > Machine Learning Processes
 - > 8 threads
- > CPU GPU Paradigm
- > Master/Slave configuration used
 - > Decentralized Management
- > Raspberry Pi
 - > Fully functioning computer
 - > Cheap \$30 Units
- > Scalability
 - > Reduces cost
- > Risks associated with non-sequential approach



Future Plans

- > Complete partnership with Dayton Law Enforcement
 - > Footprint Integration
 - > Security vs. Privacy Issues
- > Run time must be lowered
 - > Spark System
- > Continue testing complete system with high quality dataset
- > Uses existing visual sensing infrastructures
- > Train system with Haar Classifier to increase future accuracy

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- > All Photos taken from Royalty-Free Sources

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This project is dedicated to every innocent person that was killed in a drunk or distracted driving accident.